

# 5. Edge Detection, Image Segmentation and Mathematical Morphology

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- Chapter 7 (Image Segmentation)
- Chapter 8 pp.518-528 (Morphology)

# Edge Detection Using Derivative Operators

- **Edges:** the image portions which have large gradients
- Magnitude of the gradient

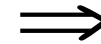
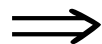
$$|\nabla f(n_1, n_2)| = \sqrt{\left[ \frac{\partial f(n_1, n_2)}{\partial n_1} \right]^2 + \left[ \frac{\partial f(n_1, n_2)}{\partial n_2} \right]^2}$$

often approximated by  $\left| \frac{\partial f(\mathbf{n}, \mathbf{n}_2)}{\partial \mathbf{n}_1} \right| + \left| \frac{\partial f(\mathbf{n}, \mathbf{n}_2)}{\partial \mathbf{n}_2} \right|$

Thresholding

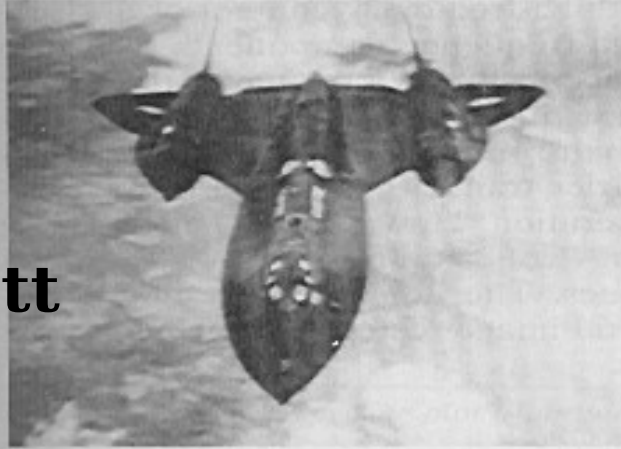
Edge Thinning & Grouping

Edge Map



# Edge Enhancement by Gradients

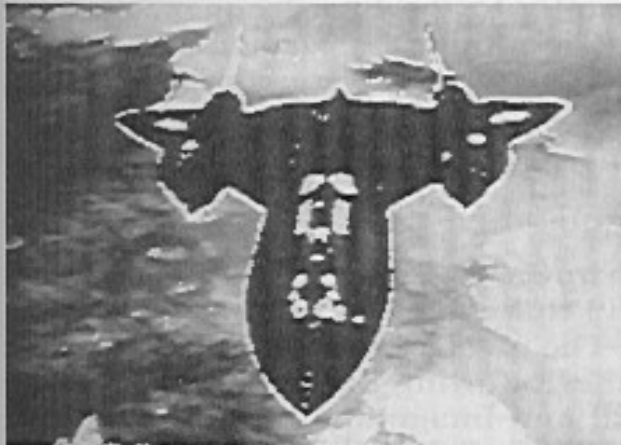
(a): original  
(b): magnitude of the gradient using the Prewitt operator  
(c): Gradient  $> 255$   $\rightarrow 255$   
(d): (c) and gradient  $< 25$   $\rightarrow 0$ .



(a)



(b)



(c)



(d)

# Edge Extraction via Gradients

(a) original

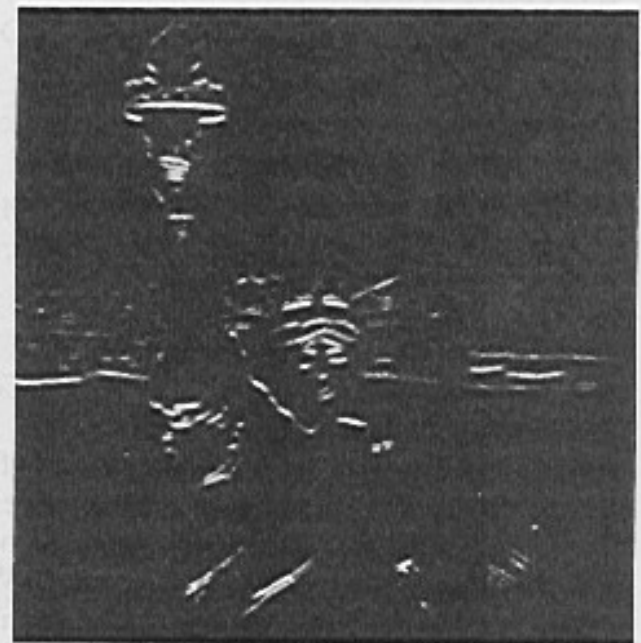
(b) vertical  
Sobel filtering

(c) horizontal  
Sobel filter

(d) magnitude of  
gradients



(a)



(b)



(c)



(d)

# Canny Edge Detector

- **Reference:** Canny, J.F., "A computational approach to edge detection," *IEEE Trans Pattern Analysis and Machine Intelligence*, 8(6): 679-698, Nov 1986.

## First Step - Smoothing

- The image is smoothed by Gaussian convolution
  - The larger the width of the Gaussian mask, the lower is the detector's sensitivity to noise.
  - The localization error in the detected edges also increases slightly as the Gaussian width is increased.

$$\frac{1}{115}$$

2	4	5	4	2
4	9	12	9	4
5	12	15	12	5
4	9	12	9	4
2	4	5	4	2

Discrete approximation to Gaussian function with  $\sigma=1.4$

# Canny Edge Detector

## Second Step—Find Edge Strength

- Find the edge strength by taking the gradient of the image
- The magnitude, or EDGE STRENGTH, of the gradient is then approximated: (Norm of the gradient)

$$|G| = |Gx| + |Gy|$$

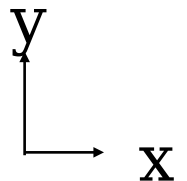
Sobel operator

-1	0	+1
-2	0	+2
-1	0	+1

Gx

+1	+2	+1
0	0	0
-1	-2	-1

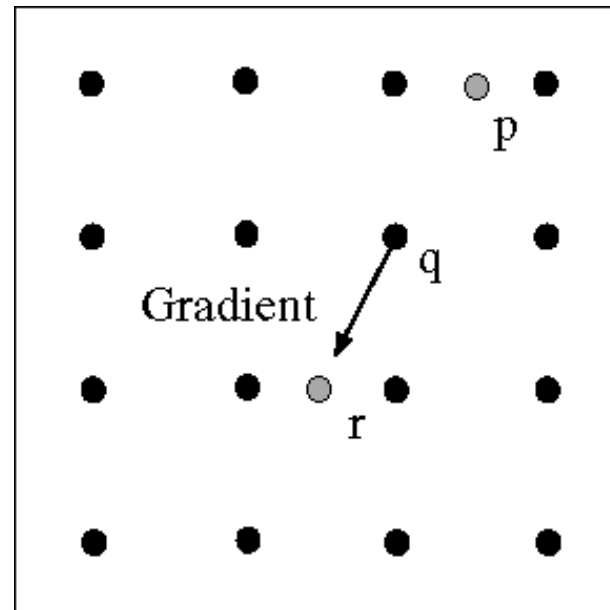
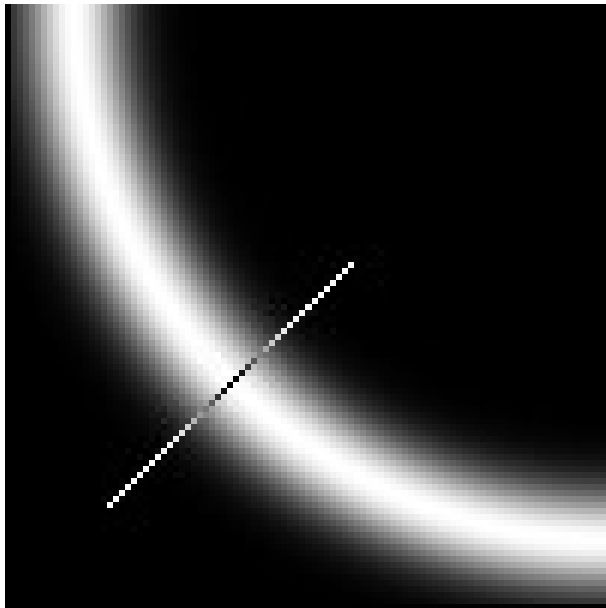
Gy



# Canny Edge Detector

## Third Step - Non-Maximum Suppression (Thinning)

Localize the peaks of the gradient magnitude



- Check if the pixel is a local maximum along the gradient direction
  - requires checking interpolated pixels p and r

# Canny Edge Detector

## Fourth Step--

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### Hysteresis

- If the threshold for the edges is too high, we have many broken lines; If too low, we have too many noise edges
- To resolve these problems, **two thresholds** are set.
- Any pixel location that has a gradient value greater than the **high threshold** is assumed to be an edge pixel.
- Then, any pixels that are connected to this edge pixel and have a gradient value greater than the **low threshold** are also selected as edge pixels.

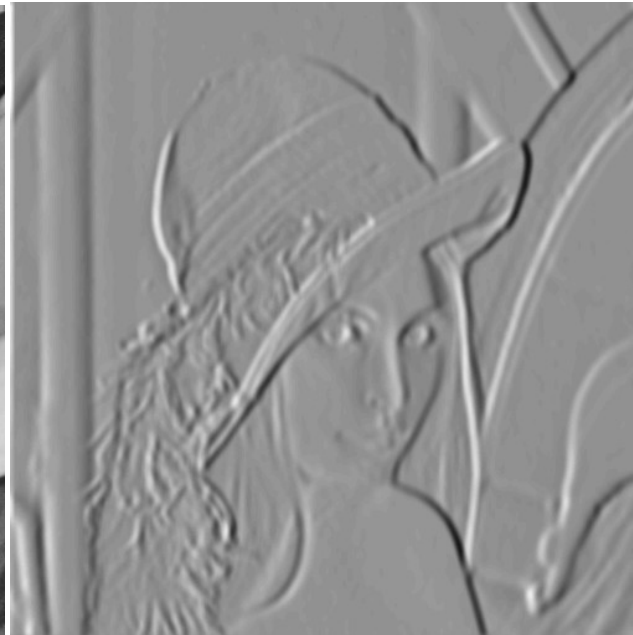


# Canny Edge Detector - An Experiment

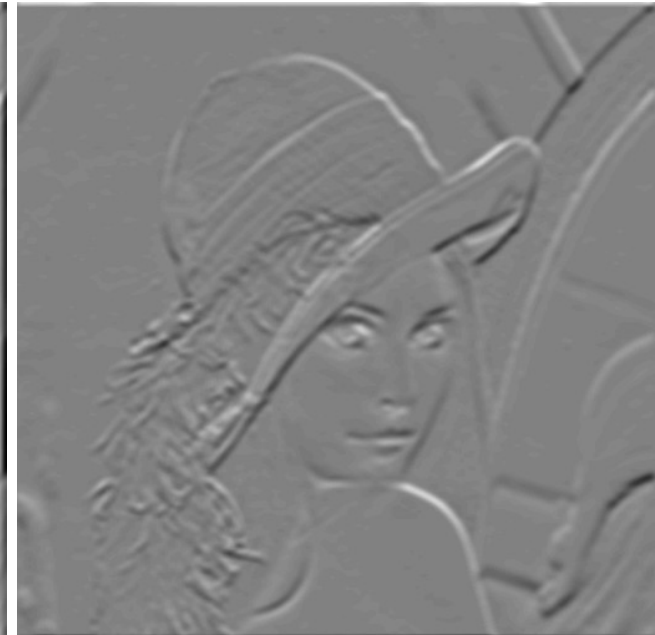
---



Original  
Image  
(Lena)



Vertical gradient



Horizontal gradient

# Final Result



Norm of the  
gradient



After thinning  
(non-maximum suppression)



After Hysteresis

# Effect of $\sigma$ (Gaussian kernel size)



Original



Canny with  $\sigma = 1$



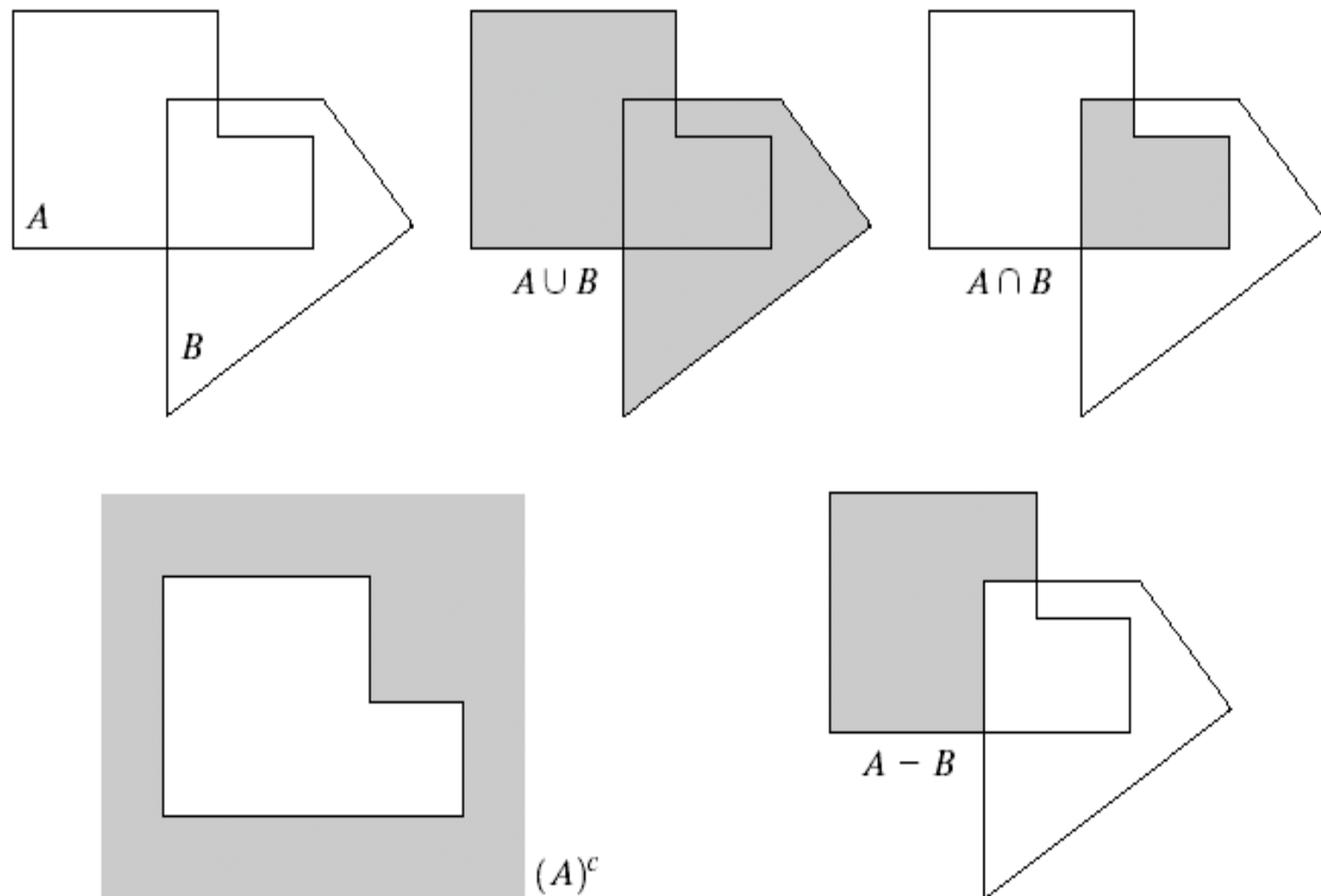
Canny with  $\sigma = 2$

- The choice of  $\sigma$  depends on desired behavior
  - large  $\sigma$  detects large scale edges
  - small  $\sigma$  detects fine features

# Mathematical Morphology

- J. Serra, **Image Analysis and Mathematical Morphology**, Academic Press, London, 1982.
- A range of non-linear image processing techniques that deal with the shape or morphology of features in an image.
- The word morphology refers to **form** and **structure**.
- Most morphological techniques operate on **binary images** for
  - Noise reduction (without eliminating essential features)
  - and feature detection
    - Analysis of connectivity of components
    - Object selection using geometric features
    - Post-processing for image segmentation

# Basic Set Theory



a	b	c
d	e	

**FIGURE 9.1**

(a) Two sets  $A$  and  $B$ . (b) The union of  $A$  and  $B$ . (c) The intersection of  $A$  and  $B$ . (d) The complement of  $A$ . (e) The difference between  $A$  and  $B$ .

# Elementary Operations

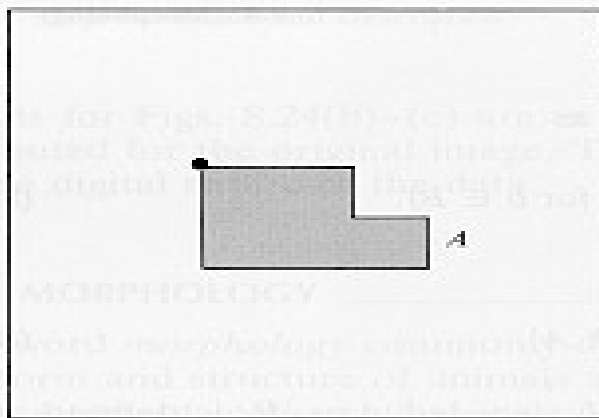
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Translation  $(A)_z = \{x/x = a + z, \text{ for } a \in A\}$

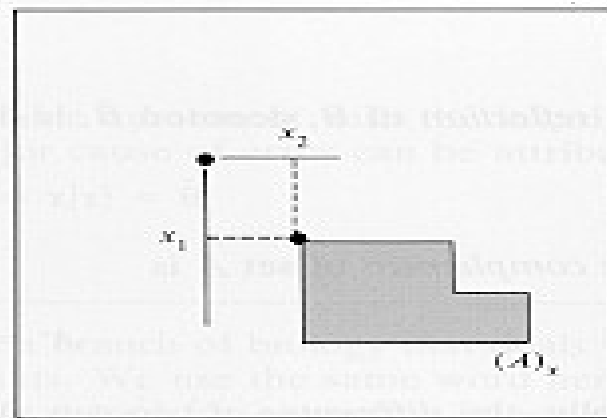
Reflection  $A^\wedge = \{x/x = -a, \text{ for } a \in A\}$

Complement  $A^c = \{x/x \notin A\}$

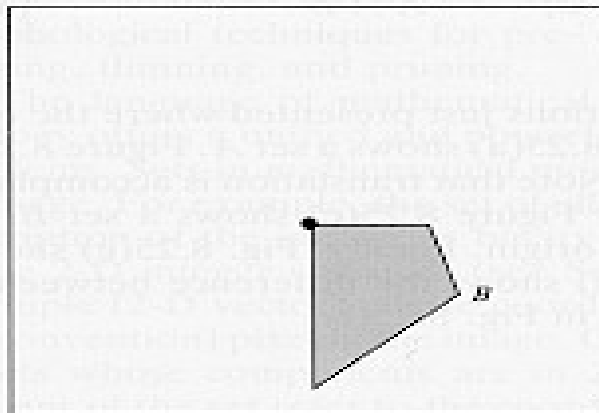
Difference  $A - B = \{x/x \in A, x \notin B\} = A \cap B^c$



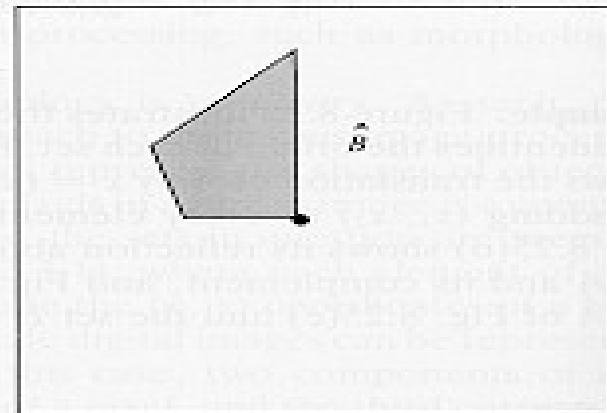
(a)



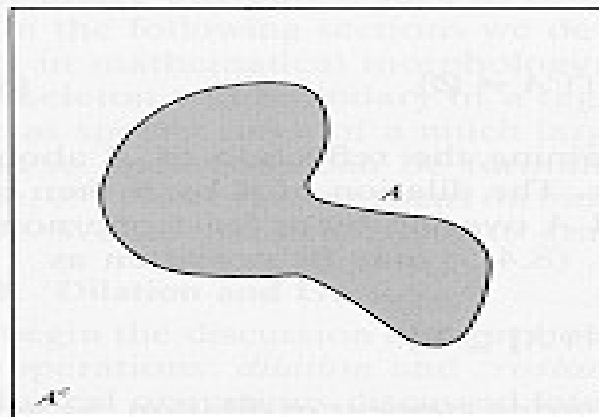
(b)



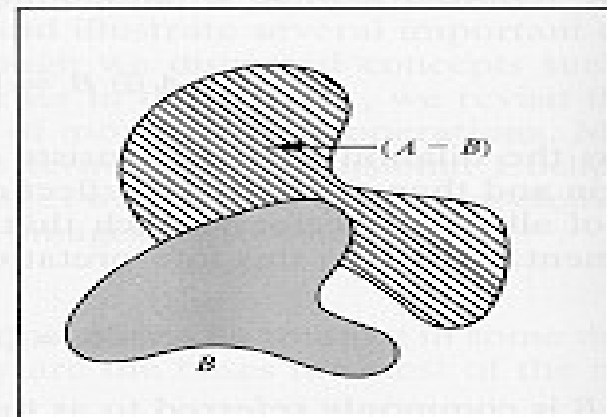
(c)



(d)



(e)



(f)

# Structuring Elements

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- Morphological operations use a small shape or template known as a **structuring element (SE)**.
- The structuring element is positioned at all possible locations in the image and is compared to the corresponding neighborhood of pixels.
- Morphological operations differ in how they carry out this comparison. Some test whether the SE “**fits**” within the neighborhood, others test whether it “**hits**” or intersects the neighborhood.



# Structuring Elements

- The structuring element applied to a binary image can be represented as a small matrix of pixels, each with a value of 1 or 0.
- The dimensions of the matrix determine the size of the SE, and its shape is determined by the pattern of ones and

$$SE = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \end{bmatrix} \quad \begin{bmatrix} 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 1 & 1 & 1 & 1 & 1 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \end{bmatrix} \quad \begin{bmatrix} 0 & 1 & 1 & 1 & 0 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 0 & 1 & 1 & 1 & 0 \end{bmatrix}$$

# Characteristics of an SE

- A structuring element has an origin (just like convolution mask).
- A SE is analogous to the kernel in convolution.
- The shape and size of the SE must be adapted to the geometric properties of the image objects of interest, i.e., an SE takes into account a number of factors: shape, size and orientation.
- Disk SE examples

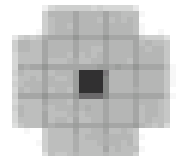
4-connected dilation



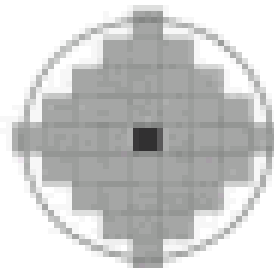
8-connected dilation



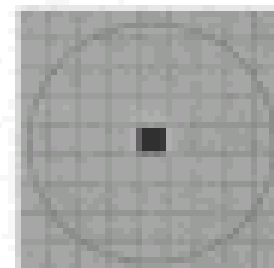
24-connected dilation



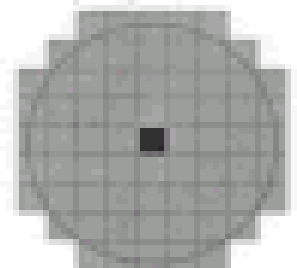
4 D4



4 D8



4 D(24)



# Basic Morphological Operations

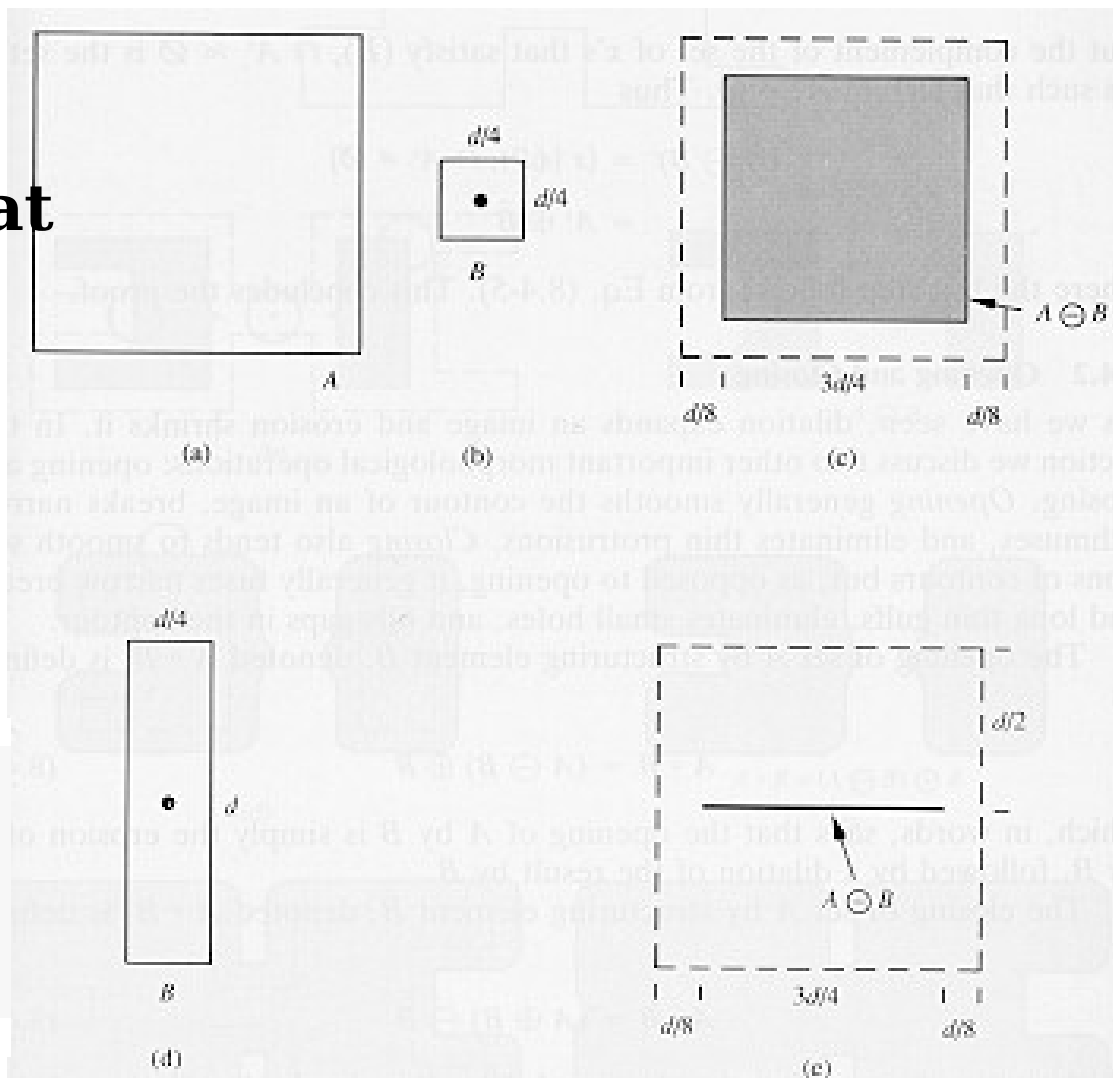
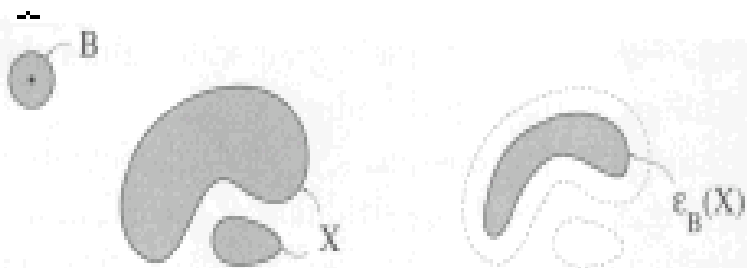
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- **Erosion:** shrinks or erodes a region:
  - It expands the holes enclosed by a single region and make the gaps between different regions larger.
- **Dilation:** expands or dilates a region
  - It shrinks the holes enclosed by a single region and make the gaps between different regions smaller.

# Erosion

The set of all  $x$  such that  $B$  translated by  $x$  is contained in  $A$ .

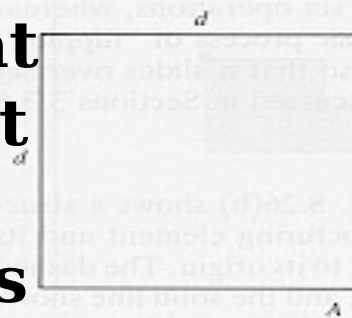
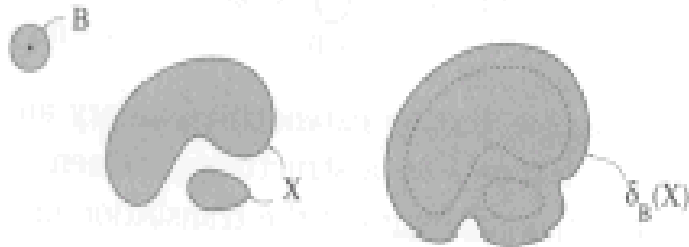
$$A \ominus B = \{x / (B_x) \subseteq A\}$$



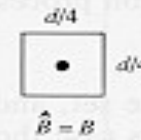
# Dilation

The set of all  $x$  such that the reflection of  $B$  about  $x$  as origin and then translated by  $x$ , overlaps with  $A$  by at least one nonzero element.

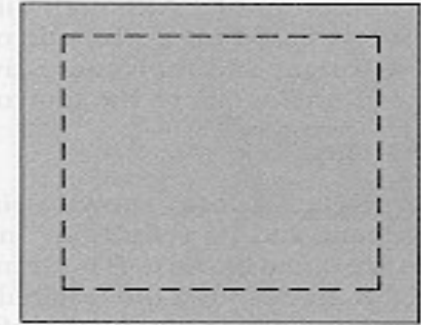
$$A \oplus B = \{x / (B)_x \cap A \neq \emptyset\}$$



(a)



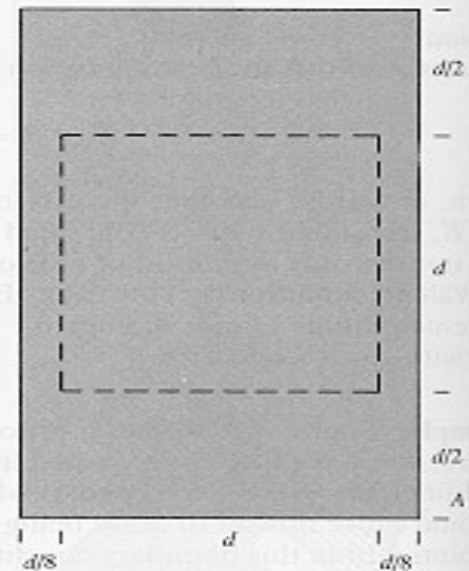
(b)



(c)



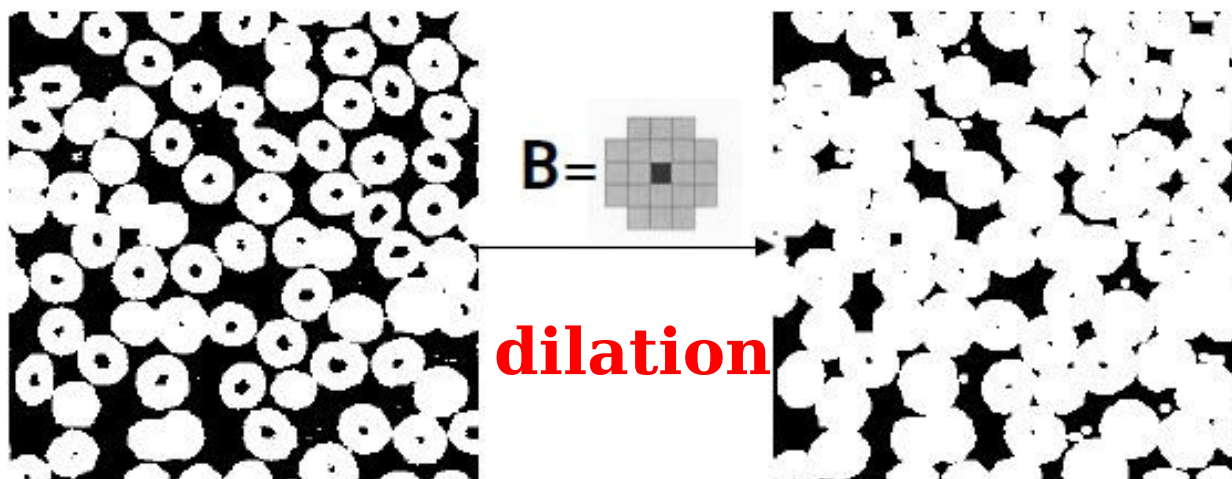
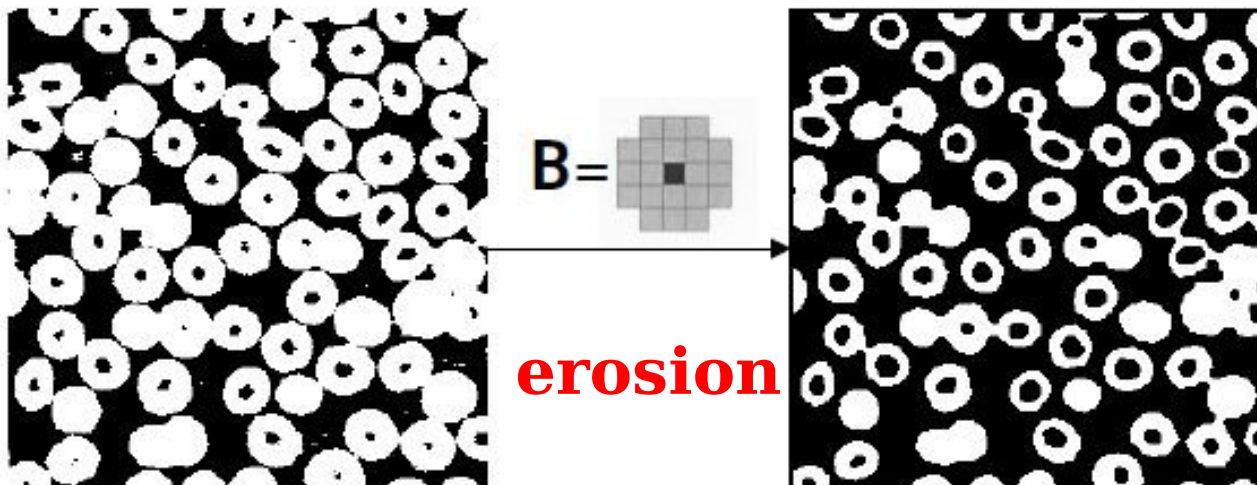
(d)



(e)

**B: structuring element**

# Erosion and Dilation



# Dilation

Historically, certain computer programs were written using only two digits rather than four to define the applicable year. Accordingly, the company's software may recognize a date using "00" as 1900 rather than the year 2000.



**Historically, certain computer programs were written using only two digits rather than four to define the applicable year. Accordingly, the company's software may recognize a date using "00" as 1900 rather than the year 2000.**



0	1	0
1	1	1
0	1	0

a c  
b

**FIGURE 9.5**

(a) Sample text of poor resolution with broken characters (magnified view).  
(b) Structuring element.  
(c) Dilation of (a) by (b). Broken segments were joined.

# Compound Morphological Operations

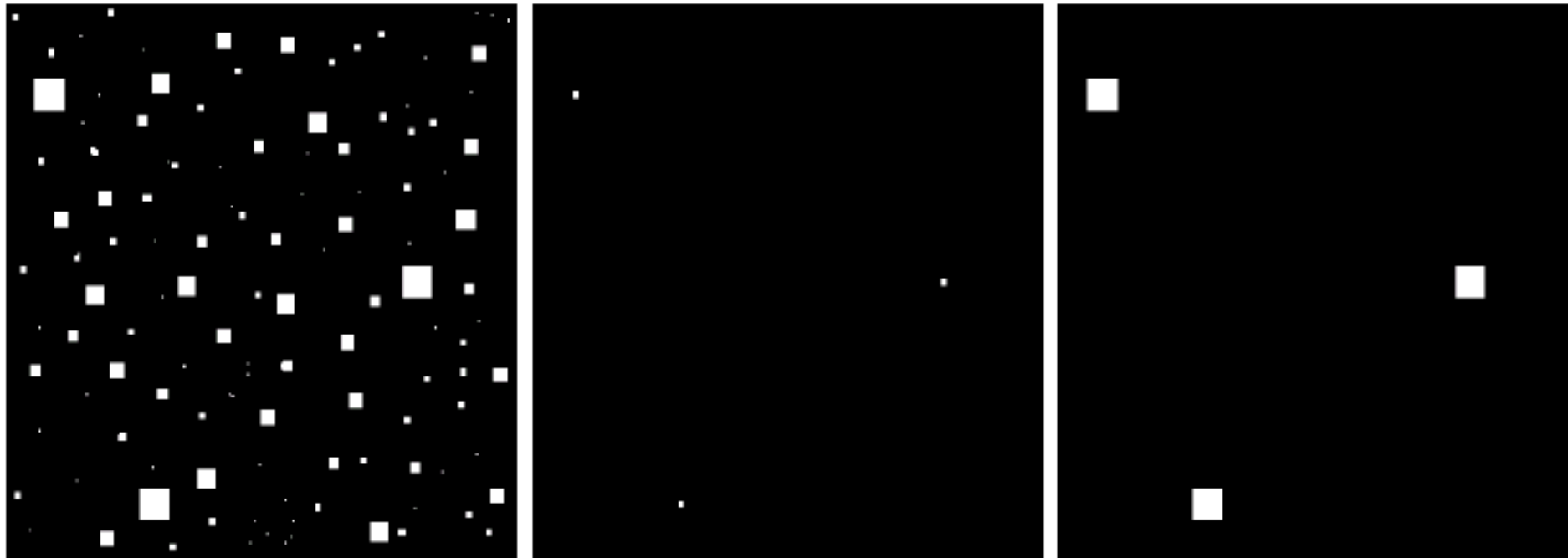
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Combinations of the elementary operations of erosion and dilation:

- **Opening:** use of erosion, followed by dilation.
  - The effect is to smooth boundaries, to break narrow isthmuses, and to eliminate small noise regions.
  - Separate connected objects, remove small objects
- **Closing:** use of dilation, followed by erosion.
  - The effect is to smooth boundaries, to join narrow breaks, and to fill small holes caused by noise.
  - Connect disconnected objects



# Opening

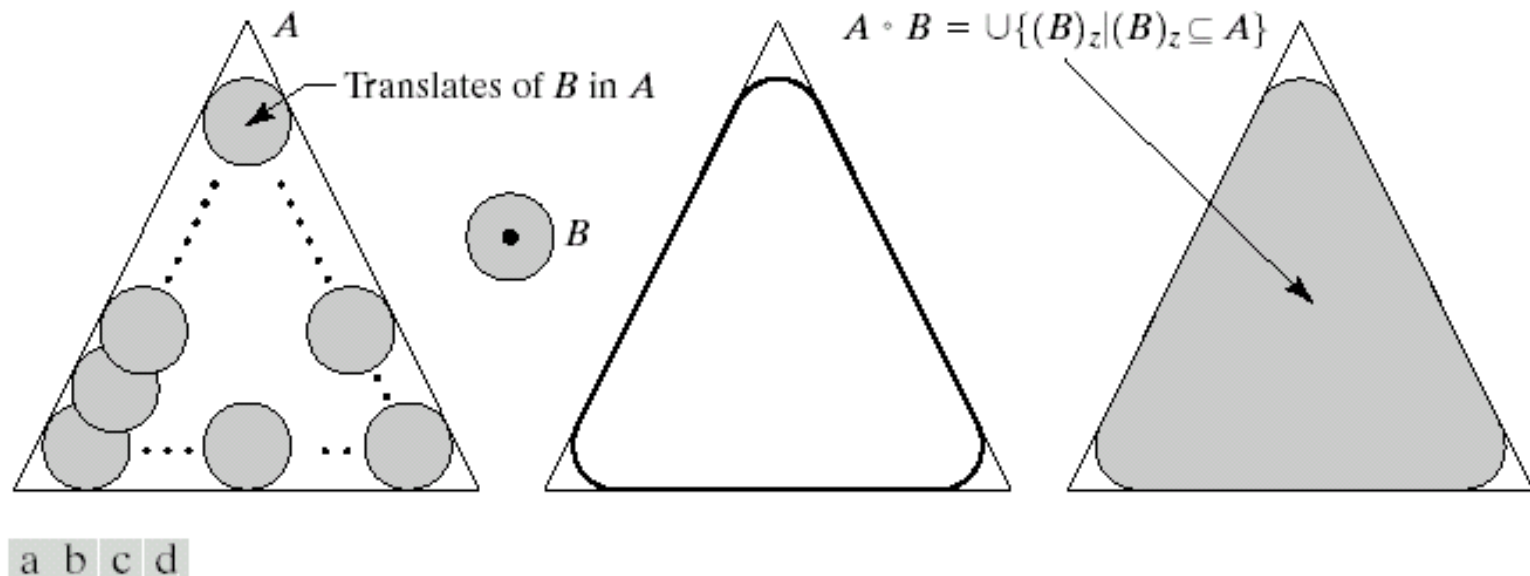


a b c

**FIGURE 9.7** (a) Image of squares of size 1, 3, 5, 7, 9, and 15 pixels on the side. (b) Erosion of (a) with a square structuring element of 1's, 13 pixels on the side. (c) Dilation of (b) with the same structuring element.

# Opening

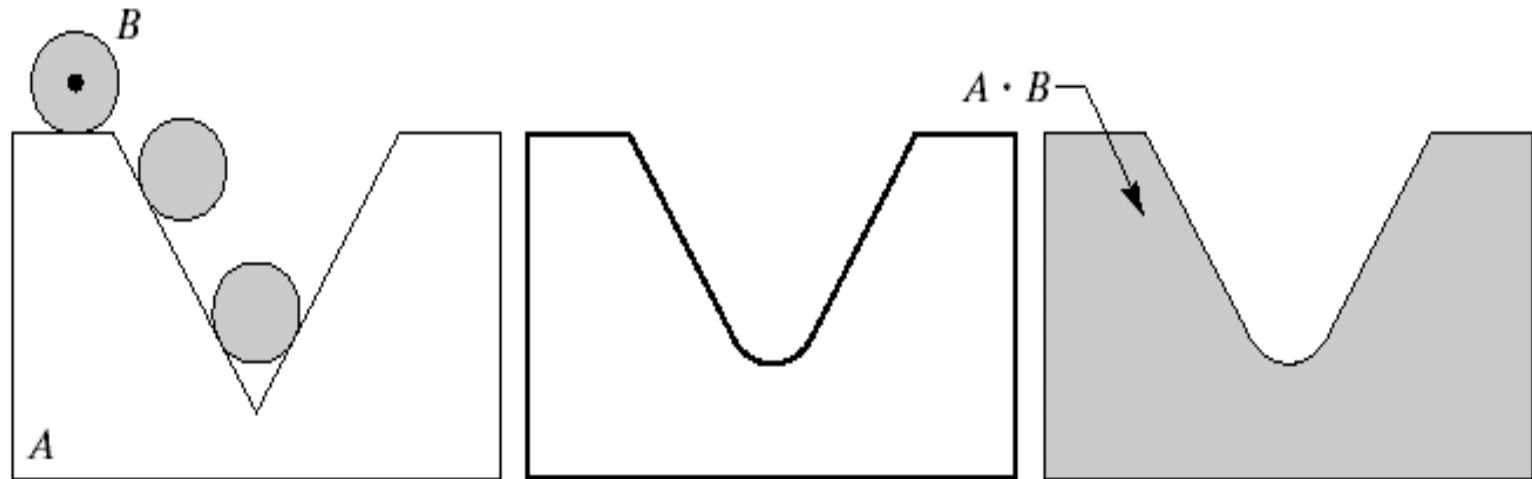
$$A \circ B = (A \ominus B) \oplus B$$



**FIGURE 9.8** (a) Structuring element  $B$  “rolling” along the inner boundary of  $A$  (the dot indicates the origin of  $B$ ). (c) The heavy line is the outer boundary of the opening. (d) Complete opening (shaded).

# Closing

$$A \bullet B = (A \oplus B) \ominus B$$



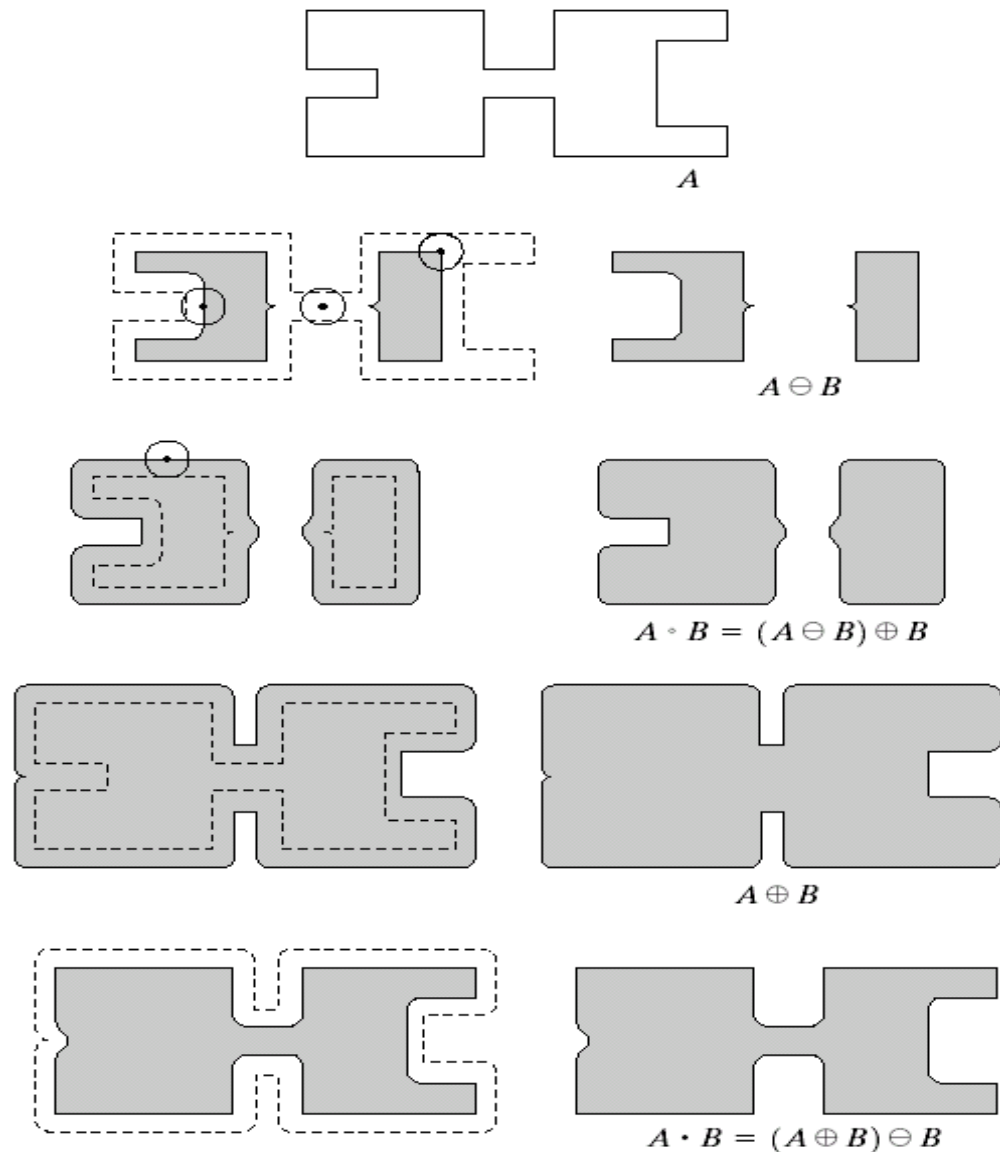
a b c

**FIGURE 9.9** (a) Structuring element  $B$  “rolling” on the outer boundary of set  $A$ . (b) Heavy line is the outer boundary of the closing. (c) Complete closing (shaded).

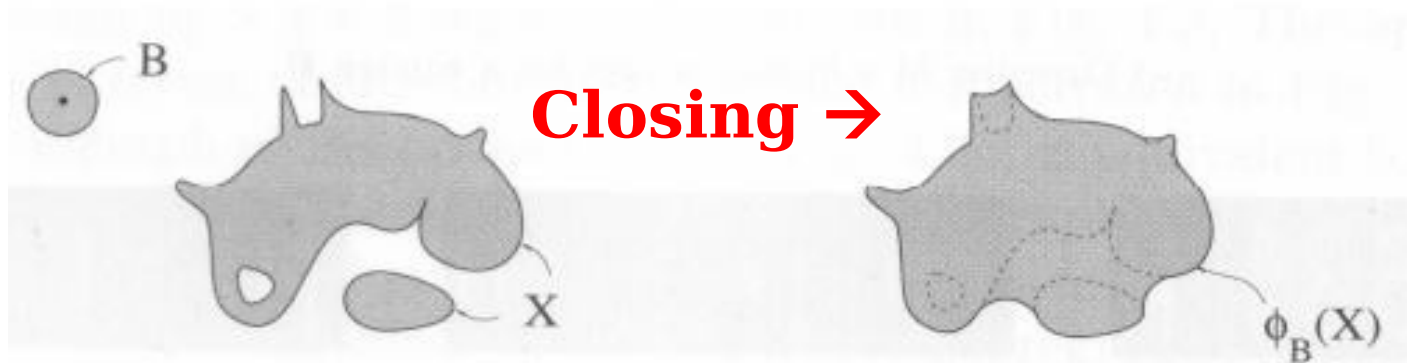
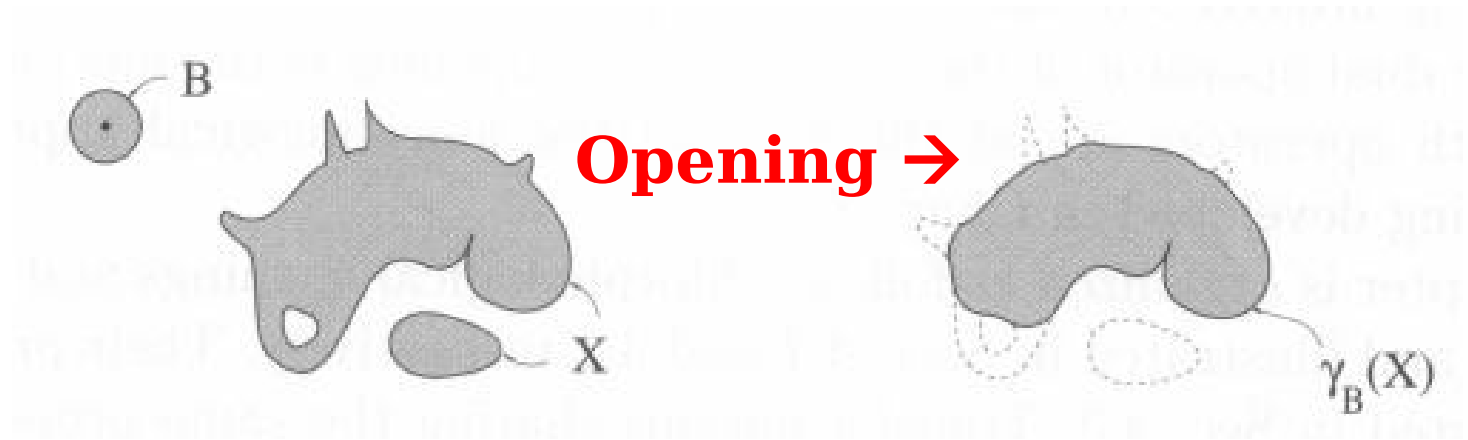
# Opening and Closing

a  
b c  
d e  
f g  
h i

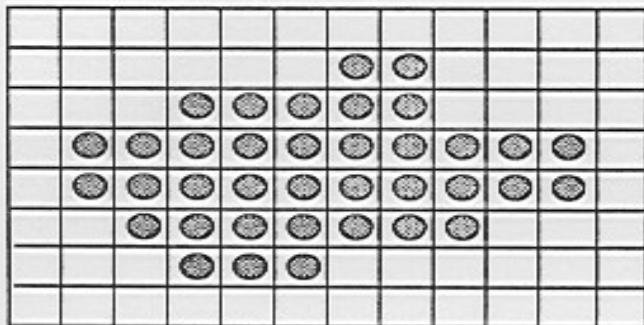
**FIGURE 9.10**  
Morphological opening and closing. The structuring element is the small circle shown in various positions in (b). The dark dot is the center of the structuring element.



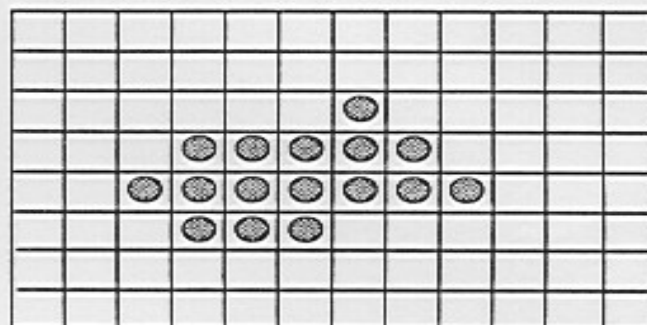
# Opening and Closing



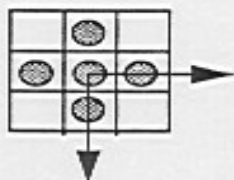
# Basic Morphological Operations



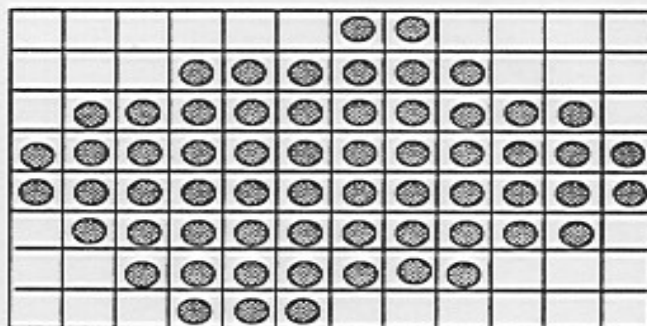
Original Binary Image: A



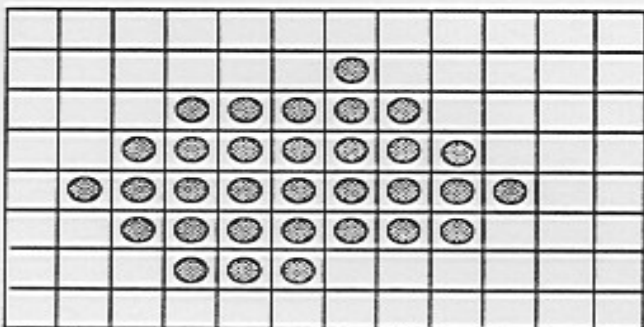
Erosion



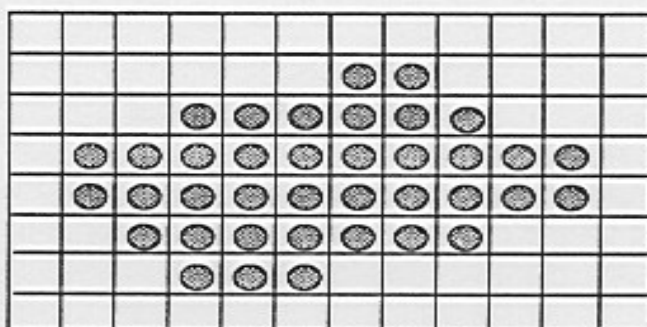
Structuring Element: K



Dilation



Opening



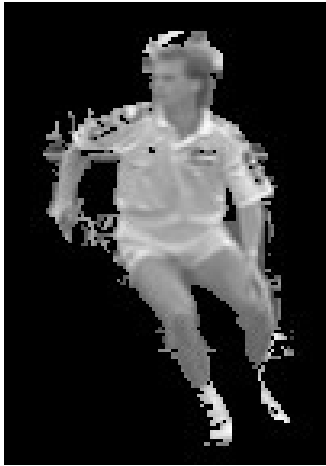
Closing

# Opening vs. Closing

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- Opening is used when the image has small regions to be removed. It is not used for narrow regions where there is a chance that the initial erosion operation might unintentionally disconnect the regions.
- Closing is used when a region has become disconnected and the desire is to restore the connectivity. It is not used when different regions are located closely and the initial dilation might connect them.

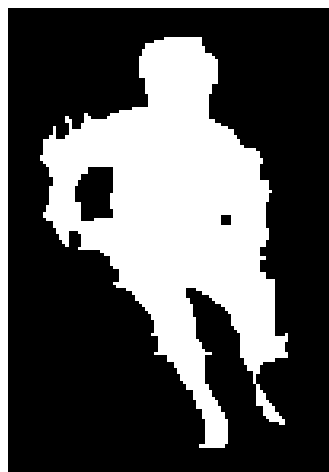
# Example (using a 3x3 structuring element)



**motion detection binary**

**erosion**

**dilation**



**opening**

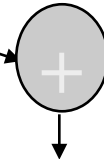
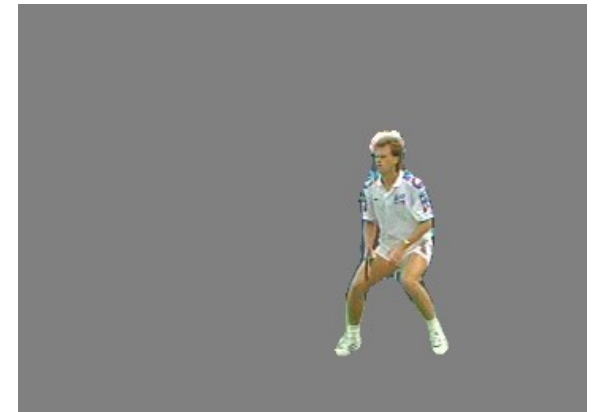
**closing**

**open-close**

**close-open**

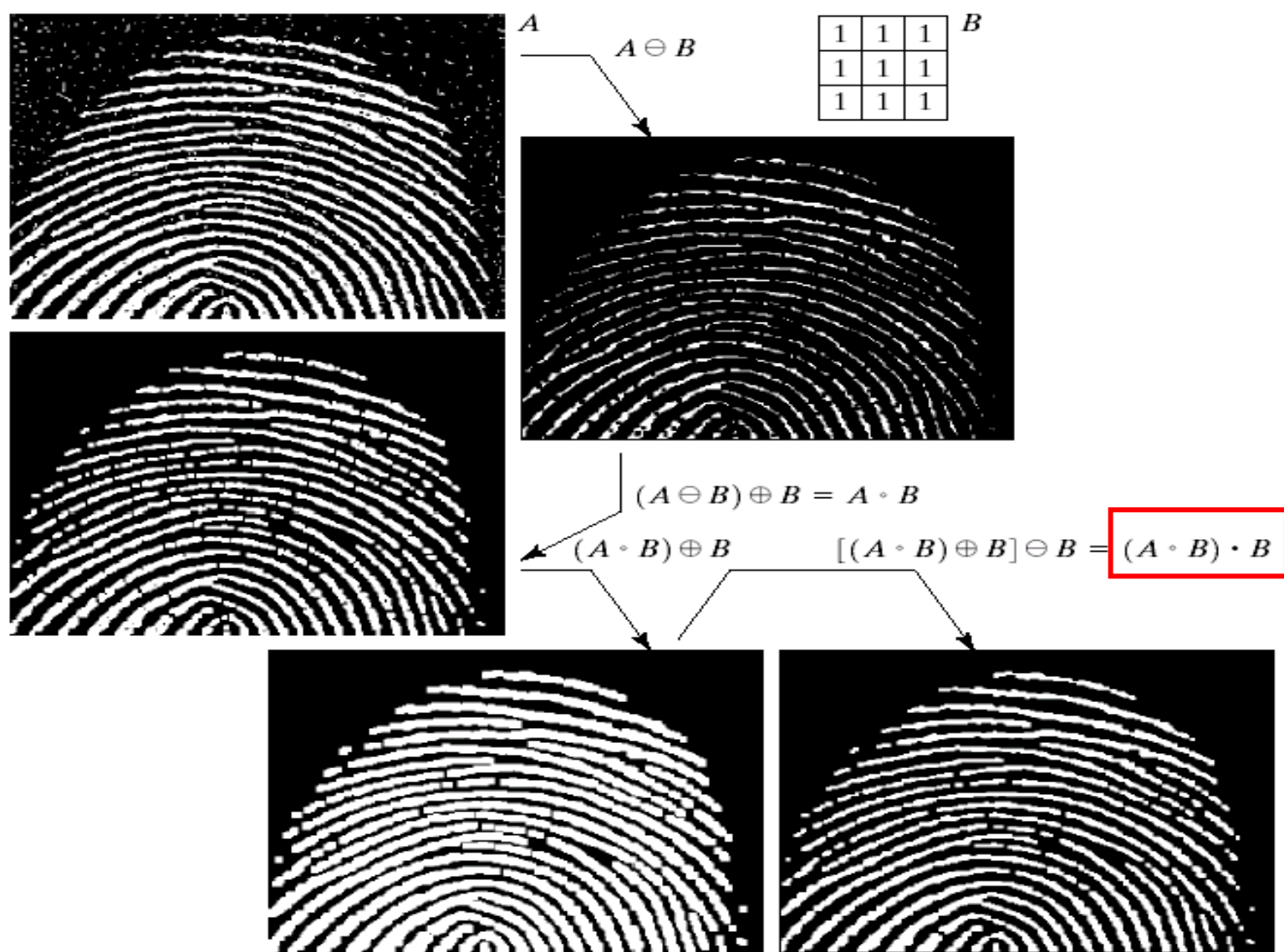


# Segmentation Application Example



**Sprite coding**

# Fingerprint Image Enhancement



**FIGURE 9.11**  
 (a) Noisy image.  
 (c) Eroded image.  
 (d) Opening of  $A$ .  
 (d) Dilated image.  
 (e) Closing of the opening. (Original image for this example courtesy of the National Institute of Standards and Technology.)

# Some Basic Morphological Algorithms

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- Hit-or-Miss Transformation
- Boundary (Contour) Extraction
- Region Filling
- Extraction of Connected Components
- Convex Hull
- Thinning
- Thickening
- Skeletonization
- Pruning (reduces short branches extruding from a region after skeletonization).

# Hit-or-Miss Transform

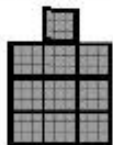
- Serves to detect features in the image that match the shape of the structuring element.
- Requires a matched pair of SEs  $\{\mathbf{B1}, \mathbf{B2}\}$  that probe the inside and outside of the feature. The hit-or-miss transform of an image  $A$  by a matched pair of SEs  $\{B_1, B_2\}$  is defined as:

$$A \circledast \{B_1, B_2\} = (A \ominus B_1) \cap (A^c \ominus B_2)$$

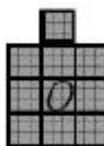
$$= (A \ominus B_1) - (A \oplus B_2)$$

# Hit-or-Miss Transform

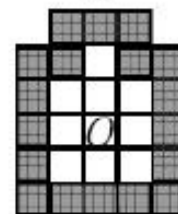
Search for:



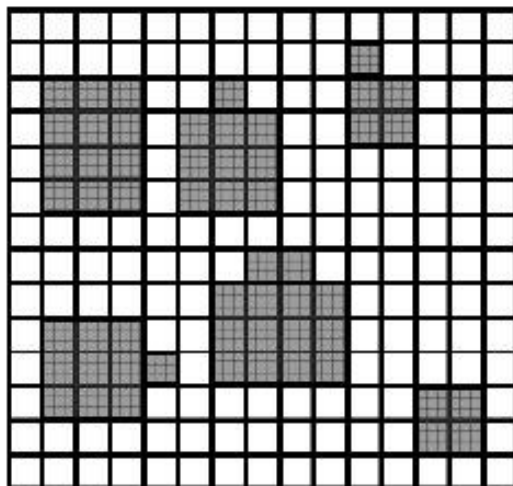
$B_1$



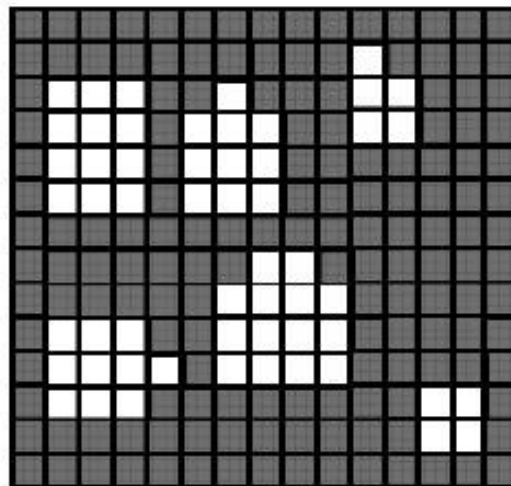
$B_2$

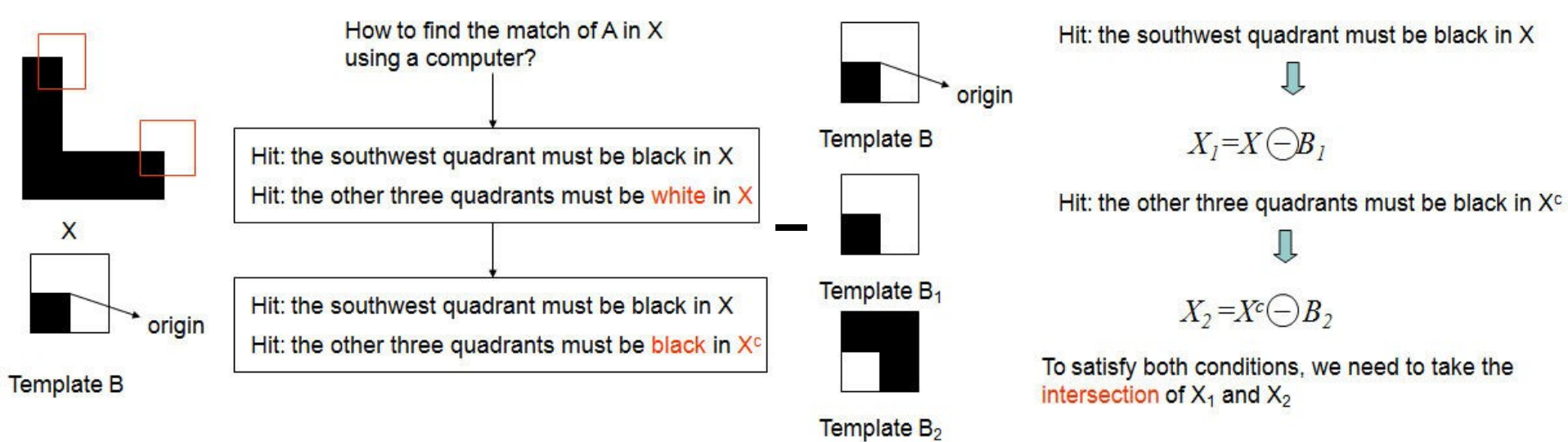


$A$

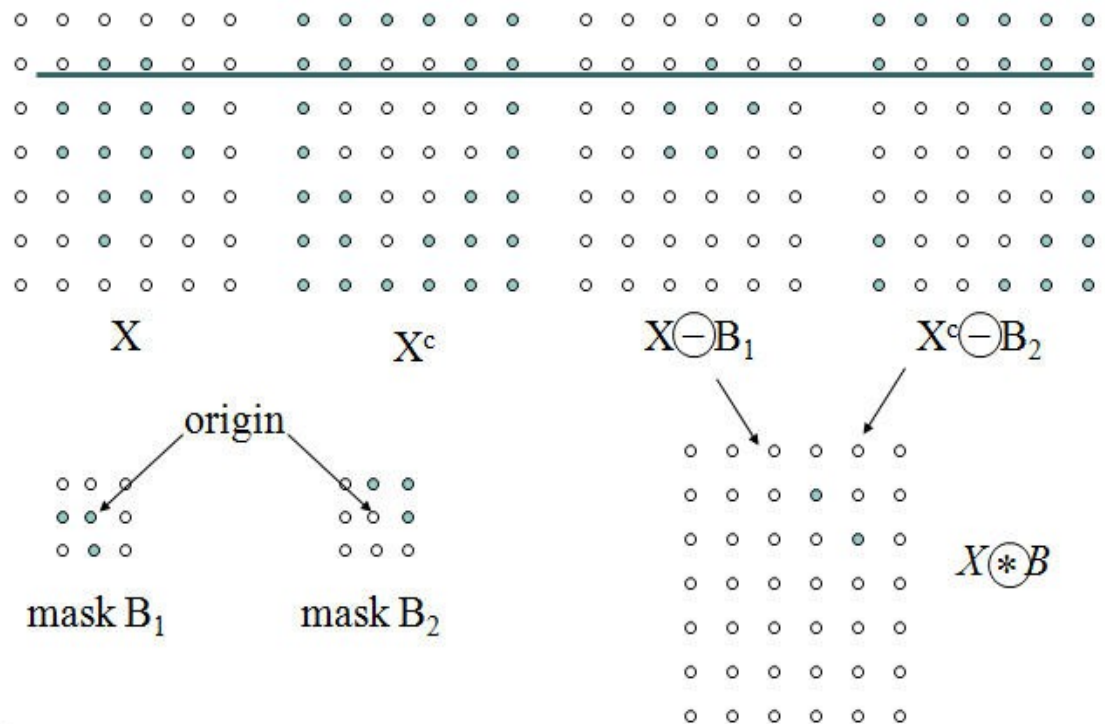


$A^c$





### Example 1



Example of  
Hit-or-Miss

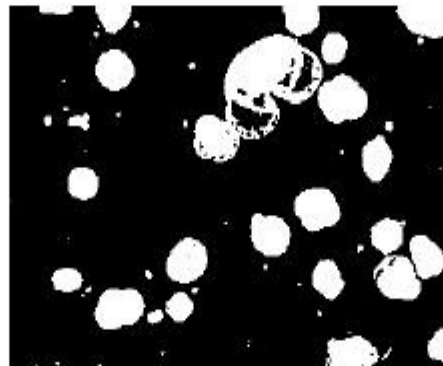
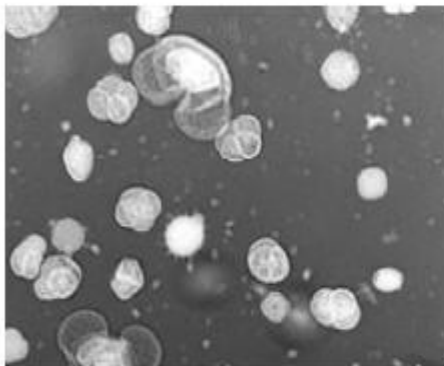
# Hit-or-Miss Transform

- Identify blobs with a radius of at least 2, but less than 4 in the pollen image. These regions totally enclose a disc of radius 2, contained in the 5 x 5 kernel named "hit" and in turn, fit within a hole of radius



CO Original Image Binary Image Objects with  $2 \leq r < 4$  l

"n



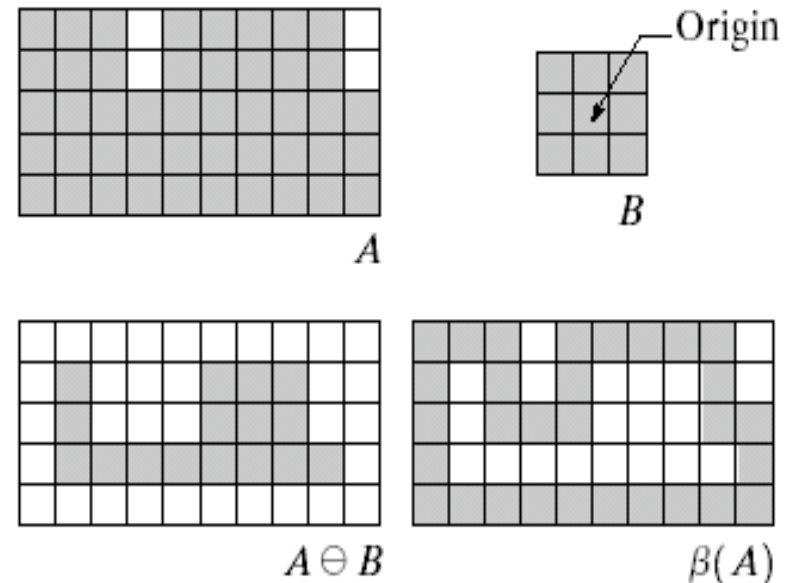


# Boundary (Contour) Extraction

- One application of erosion is contour extraction – the contours are extracted by subtracting the eroded image from the original image. The formula for this is  $\beta(A) = A - (A \ominus B)$

a b  
c d

**FIGURE 9.13** (a) Set  $A$ . (b) Structuring element  $B$ . (c)  $A$  eroded by  $B$ . (d) Boundary, given by the set difference between  $A$  and its erosion.





# Boundary Extraction Example



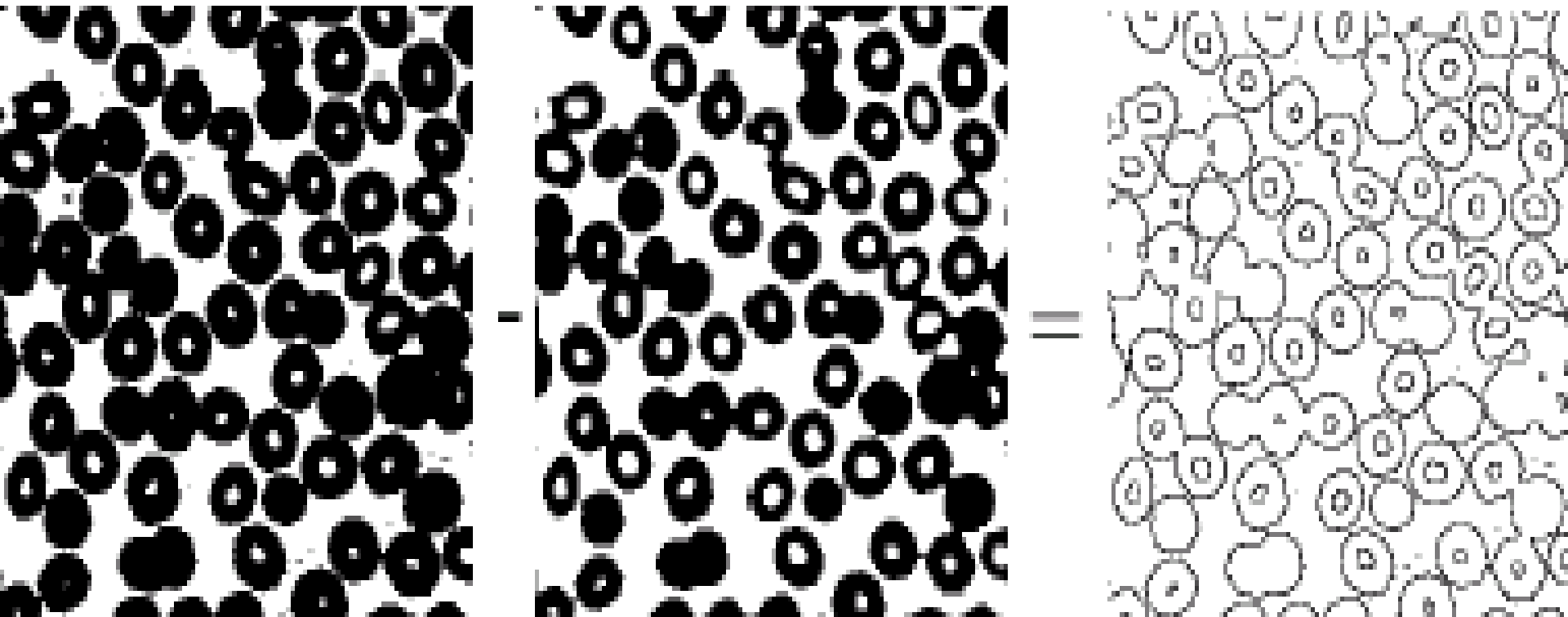
a b

**FIGURE 9.14**

(a) A simple binary image, with 1's represented in white. (b) Result of using Eq. (9.5-1) with the structuring element in Fig. 9.13(b).

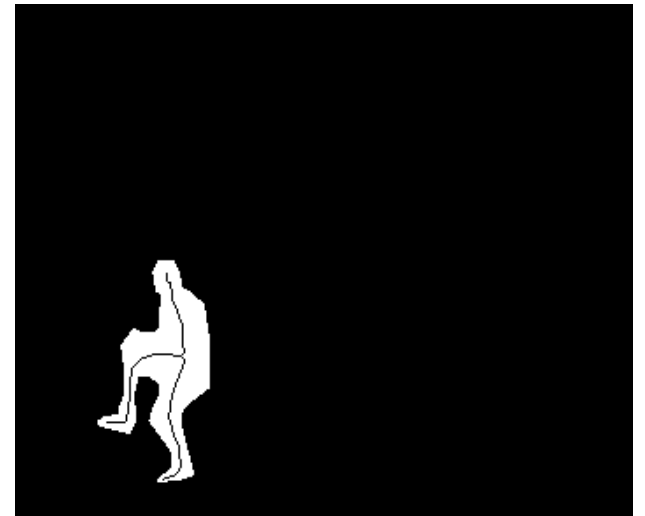
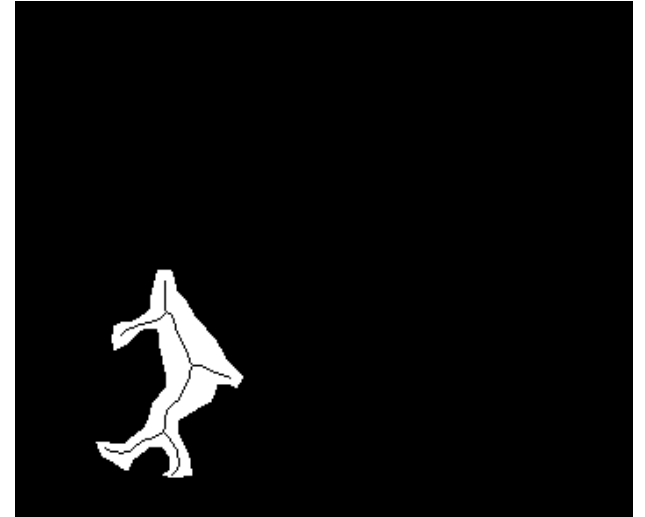
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# Boundary Extraction Example



# Application of Skeletonization

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**TABLE 9.2**  
Summary of  
morphological  
operations and  
their properties.

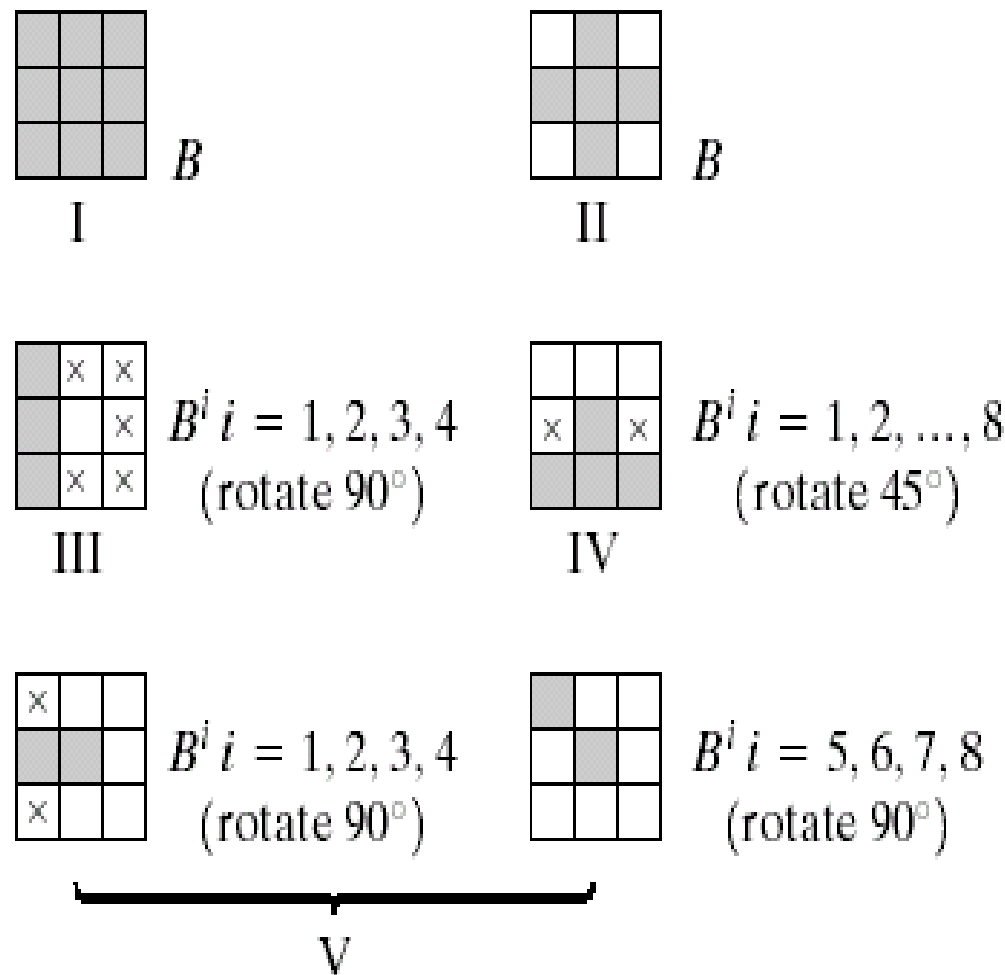
		Comments (The Roman numerals refer to the structuring elements shown in Fig. 9.26).
Operation	Equation	
Translation	$(A)_z = \{w   w = a + z, \text{ for } a \in A\}$	Translates the origin of $A$ to point $z$ .
Reflection	$\hat{B} = \{w   w = -b, \text{ for } b \in B\}$	Reflects all elements of $B$ about the origin of this set.
Complement	$A^c = \{w   w \notin A\}$	Set of points not in $A$ .
Difference	$A - B = \{w   w \in A, w \notin B\}$ $= A \cap B^c$	Set of points that belong to $A$ but not to $B$ .
Dilation	$A \oplus B = \{z   (\hat{B})_z \cap A \neq \emptyset\}$	“Expands” the boundary of $A$ . (I)
Erosion	$A \ominus B = \{z   (B)_z \subseteq A\}$	“Contracts” the boundary of $A$ . (I)
Opening	$A \circ B = (A \ominus B) \oplus B$	Smooths contours, breaks narrow isthmuses, and eliminates small islands and sharp peaks. (I)
Closing	$A \bullet B = (A \oplus B) \ominus B$	Smooths contours, fuses narrow breaks and long thin gulfs, and eliminates small holes. (I)

Hit-or-miss transform	$A \circledast B = (A \ominus B_1) \cap (A^c \ominus B_2)$ $= (A \ominus B_1) - (A \oplus \hat{B}_2)$	The set of points (coordinates) at which, simultaneously, $B_1$ found a match (“hit”) in $A$ and $B_2$ found a match in $A^c$ .
Boundary extraction	$\beta(A) = A - (A \ominus B)$	Set of points on the boundary of set $A$ . (I)
Region filling	$X_k = (X_{k-1} \oplus B) \cap A^c; X_0 = p \text{ and } k = 1, 2, 3, \dots$	Fills a region in $A$ , given a point $p$ in the region. (II)
Connected components	$X_k = (X_{k-1} \oplus B) \cap A; X_0 = p \text{ and } k = 1, 2, 3, \dots$	Finds a connected component $Y$ in $A$ , given a point $p$ in $Y$ . (I)
Convex hull	$X_k^i = (X_{k-1}^i \circledast B^i) \cup A; i = 1, 2, 3, 4;$ $k = 1, 2, 3, \dots; X_0^i = A; \text{ and}$ $D^i = X_{\text{conv}}^i.$	Finds the convex hull $C(A)$ of set $A$ , where “conv” indicates convergence in the sense that $X_k^i = X_{k-1}^i$ . (III)

**TABLE 9.2**  
Summary of  
morphological  
results and their  
properties.  
*(continued)*

Operation	Equation	Comments (The Roman numerals refer to the structuring elements shown in Fig. 9.26).
Thinning	$A \otimes B = A - (A \circledast B)$ $= A \cap (A \circledast B)^c$ $A \otimes \{B\} =$ $((\dots ((A \otimes B^1) \otimes B^2) \dots) \otimes B^n)$ $\{B\} = \{B^1, B^2, B^3, \dots, B^n\}$	<p>Thins set <math>A</math>. The first two equations give the basic definition of thinning.</p> <p>The last two equations denote thinning by a sequence of structuring elements. This method is normally used in practice. (IV)</p>
Thickening	$A \odot B = A \cup (A \circledast B)$ $A \odot \{B\} =$ $((\dots (A \odot B^1) \odot B^2 \dots) \odot B^n)$	<p>Thickens set <math>A</math>. (See preceding comments on sequences of structuring elements.) Uses IV with 0's and 1's reversed.</p>

Skeletons	$S(A) = \bigcup_{k=0} S_k(A)$ $S_k(A) = \bigcup_{k=0}^K \{ (A \ominus kB) - [(A \ominus kB) \circ B] \}$ <p>Reconstruction of <math>A</math>:</p> $A = \bigcup_{k=0}^K (S_k(A) \oplus kB)$	<p>Finds the skeleton <math>S(A)</math> of set <math>A</math>. The last equation indicates that <math>A</math> can be reconstructed from its skeleton subsets <math>S_k(A)</math>. In all three equations, <math>K</math> is the value of the iterative step after which the set <math>A</math> erodes to the empty set. The notation <math>(A \ominus kB)</math> denotes the <math>k</math>th iteration of successive erosion of <math>A</math> by <math>B</math>. (I)</p>
Pruning	$X_1 = A \otimes \{B\}$ $X_2 = \bigcup_{k=1}^8 (X_1 \circledast B^k)$ $X_3 = (X_2 \oplus H) \cap A$ $X_4 = X_1 \cup X_3$	<p><math>X_4</math> is the result of pruning set <math>A</math>. The number of times that the first equation is applied to obtain <math>X_1</math> must be specified. Structuring elements <math>V</math> are used for the first two equations. In the third equation <math>H</math> denotes structuring element <math>I</math>.</p>



**FIGURE 9.26** Five basic types of structuring elements used for binary morphology. The origin of each element is at its center and the  $\times$ 's indicate "don't care" values.